



# Towards Understanding Thermal Runaway of Lithium Batteries



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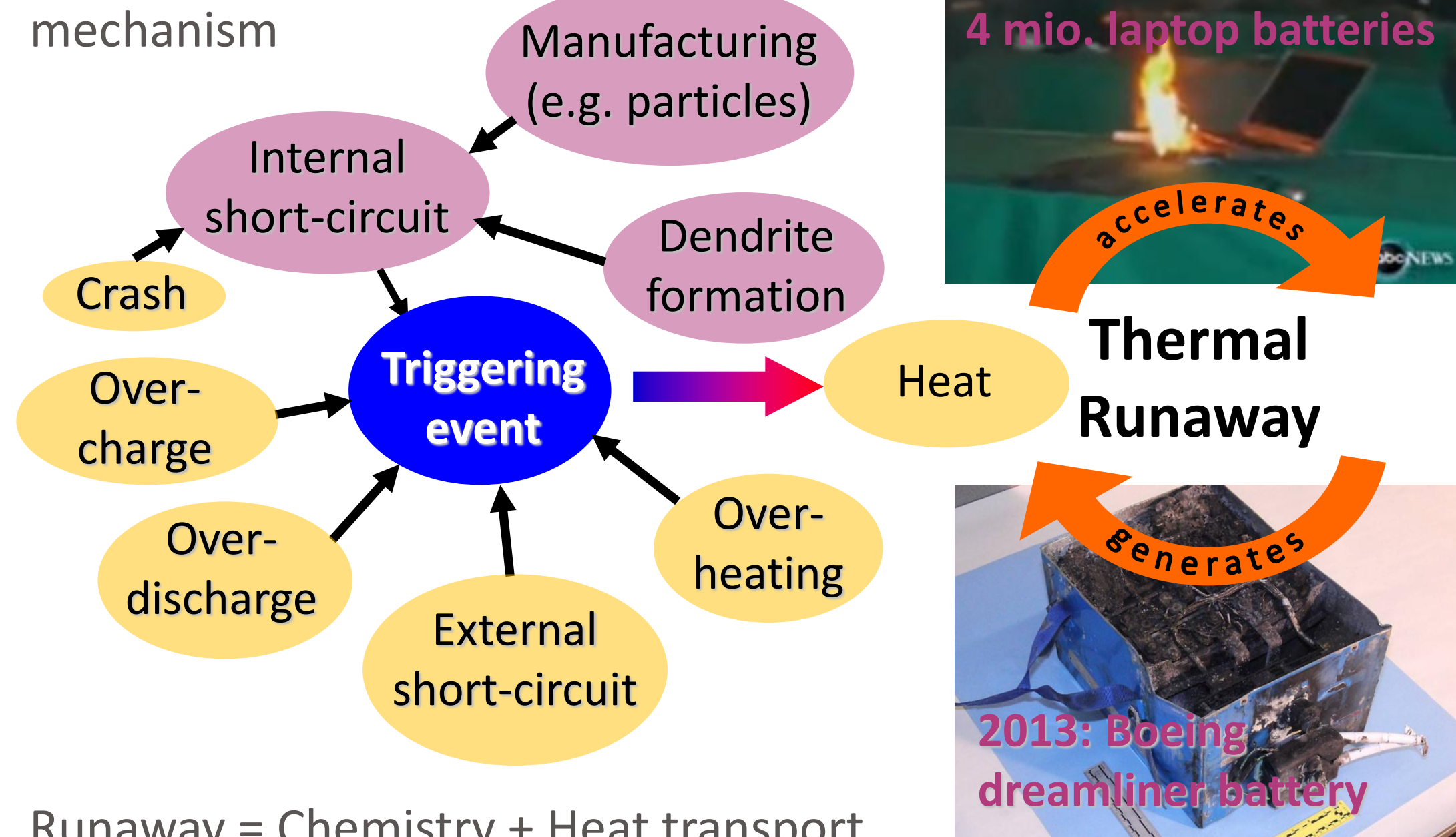
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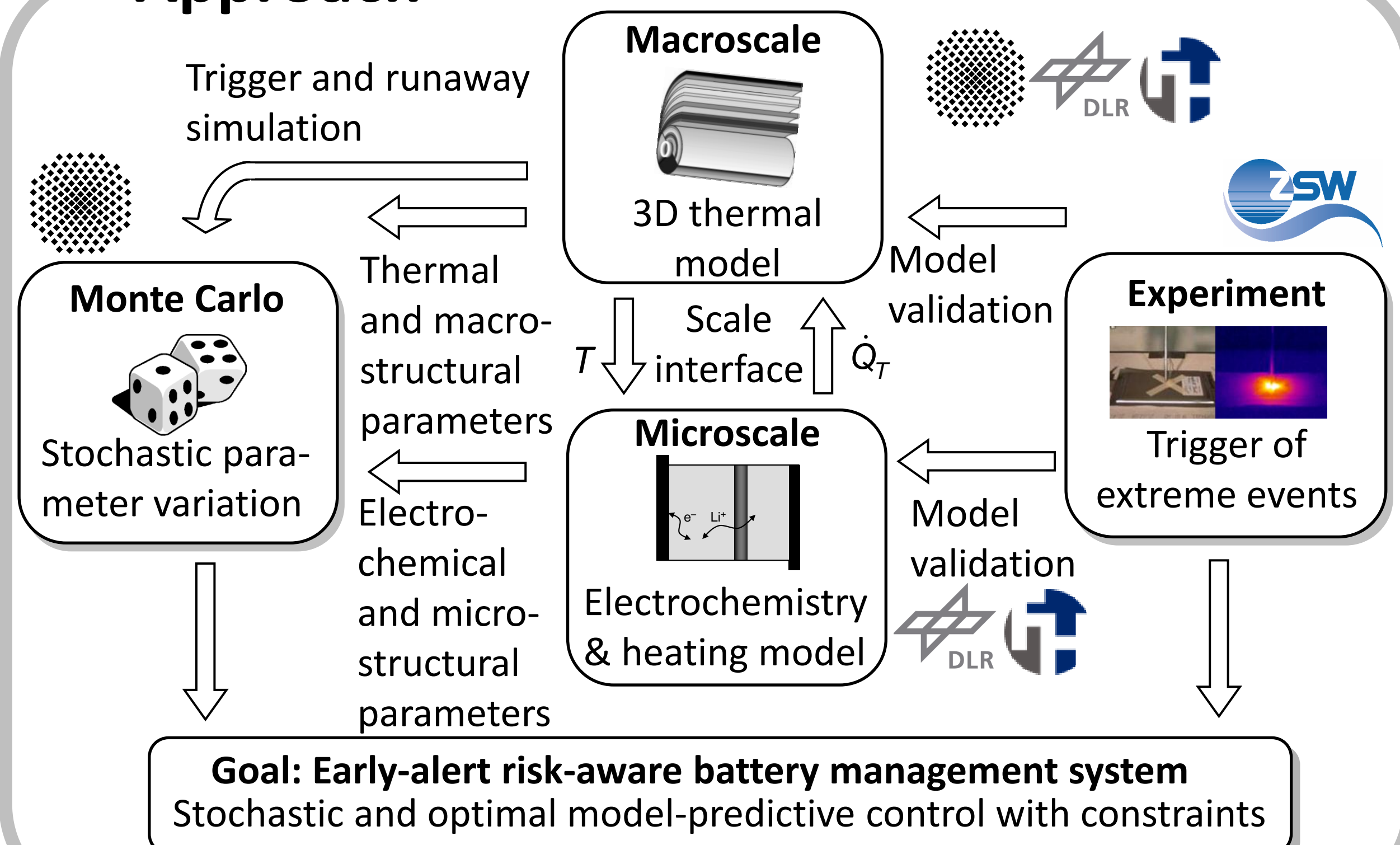
## Motivation

Thermal runaway mechanism



Runaway = Chemistry + Heat transport

## Approach



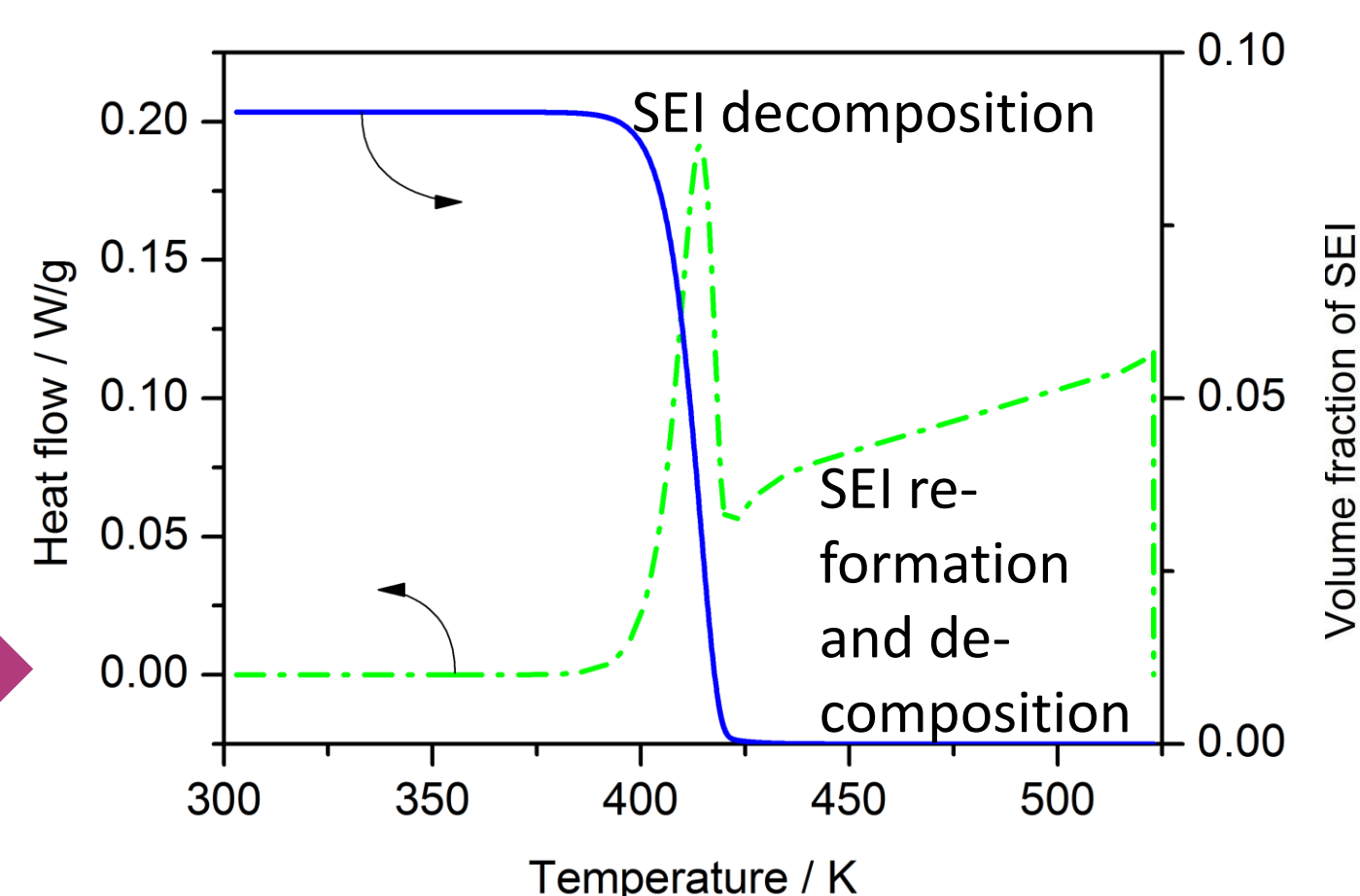
**Goal: Early-alert risk-aware battery management system**  
Stochastic and optimal model-predictive control with constraints

## Micro Model

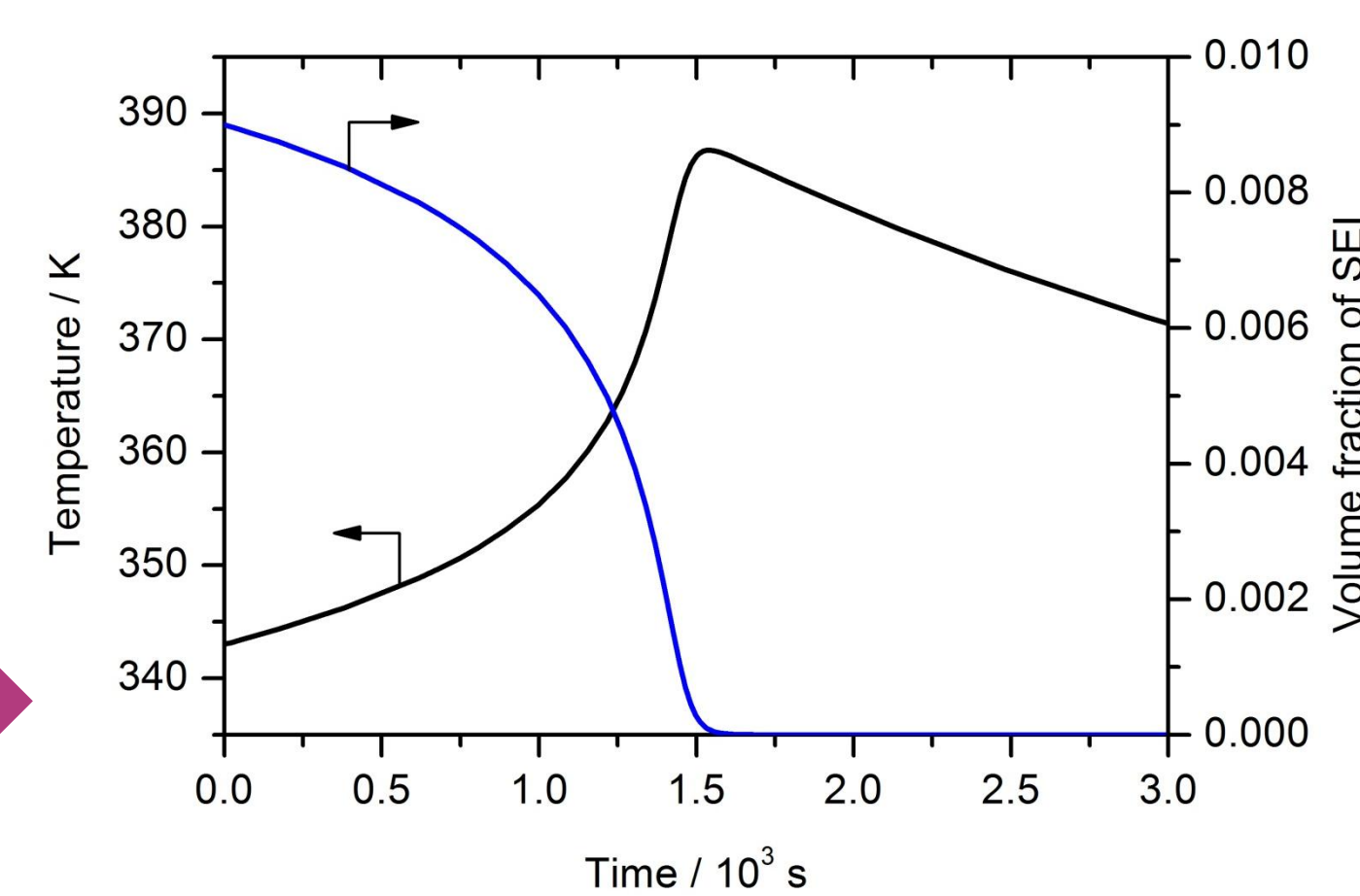
Degradation models at high temperature include:

- **Solid electrolyte interface (SEI) decomposition**  
 $(\text{CH}_2\text{OCO}_2\text{Li})_2 \rightarrow \text{Li}_2\text{CO}_3 + \text{C}_2\text{H}_4 + \text{CO}_2 + 0.5 \text{O}_2$
- **SEI formation (Electrolyte decomposition)**  
 $2 \text{C}_3\text{H}_4\text{O}_3 (\text{EC}) + 2 \text{e}^- + 2 \text{Li}^+ \rightarrow (\text{CH}_2\text{OCO}_2\text{Li})_2 + \text{C}_2\text{H}_4$
- **Electrolyte evaporation**  
 $\text{C}_3\text{H}_4\text{O}_3 (\text{liquid}) \rightarrow \text{C}_3\text{H}_4\text{O}_3 (\text{gas})$

Simulation of **differential scanning calorimetry (DSC)** for SEI decomposition and formation. Heat rate is 5K/min.



**Thermal runaway** is successfully observed at extreme condition under low heat transfer coefficient and thermal conductivity.



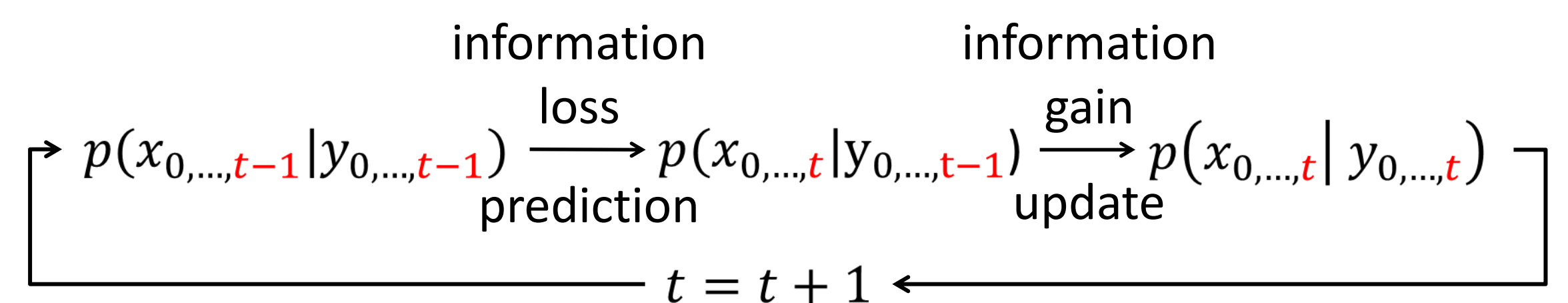
## Stochastic Model

**Bayesian filtering**

System model:  $x_t = f(x_{t-1}, \mu_t)$   $x_t$  ... model state at time  $t$   
Measurement model:  $y_t = g(x_t, v_t)$   $y_t$  ... measurement at time  $t$   
 $\mu_t$  ... model error  
 $v_t$  ... measurement error  
Update of uncertain model predictions with measurements via Bayes' theorem:

$$p(x_t | y_0, \dots, y_t) = \frac{p(y_0, \dots, y_t | x_t) \cdot p(x_t)}{p(y_0, \dots, y_t)}$$

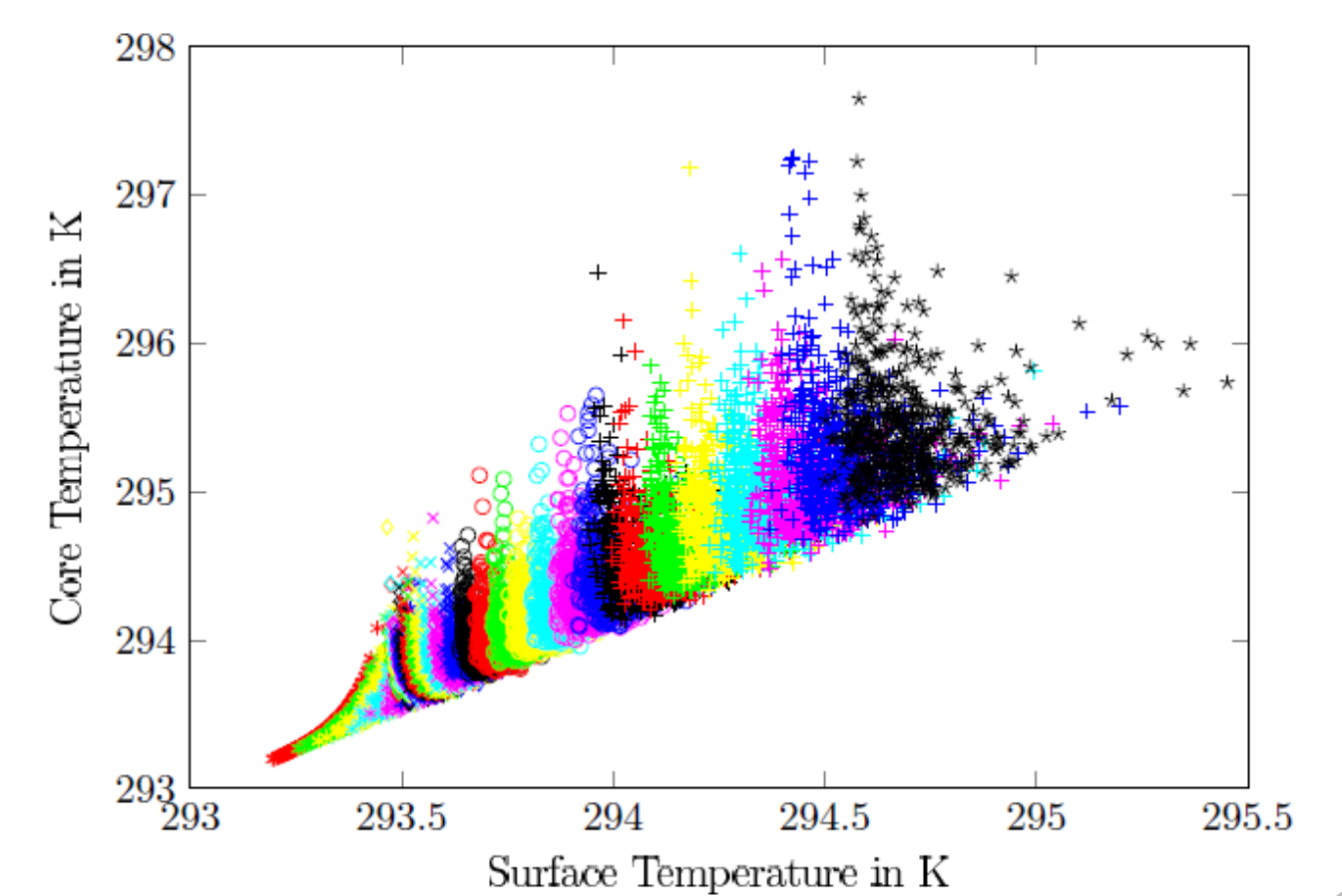
Complete sequential procedure (simplified):



Solution of model equations with a particle filter:

- Continuous probability density is discretized by particles (individual model runs)
- Measurement update via reweighting of the particles

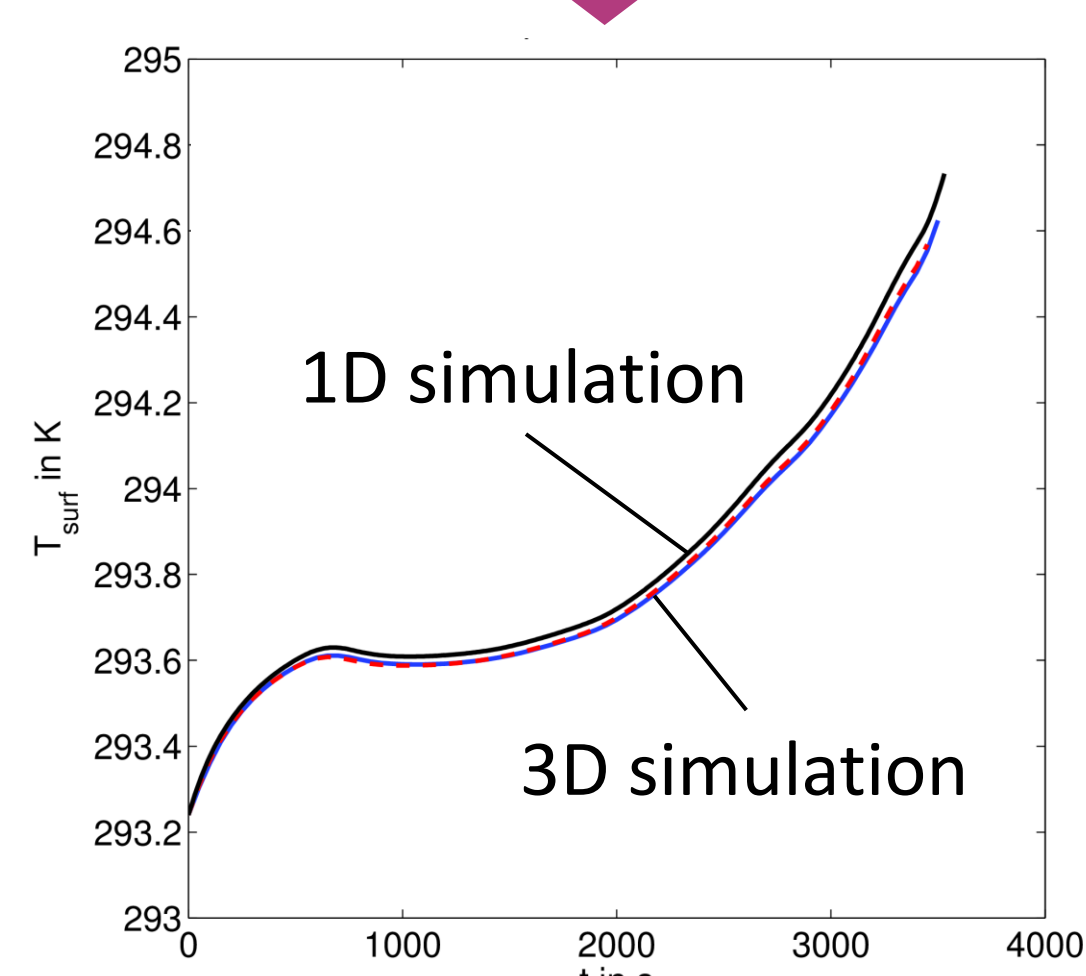
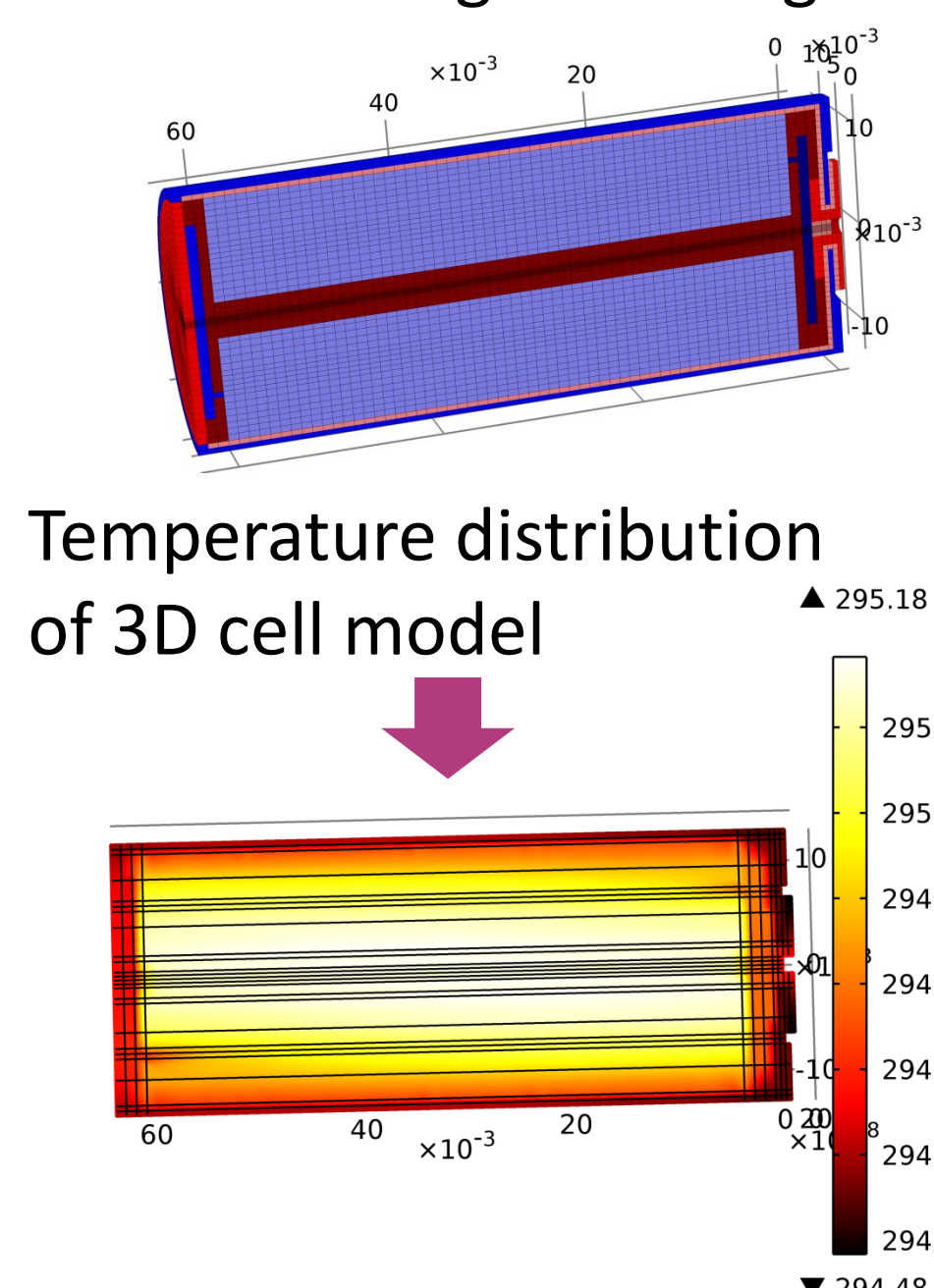
Prediction of cell core temperature as a function of surface temperature for a number of simulation runs under uncertainty



## Macro Model

3D, 2D and 1D model of single cell will be investigated using COMSOL

3D simulation is compared with 1D simulation under nominal discharge operation in 1 hour (1C rate).



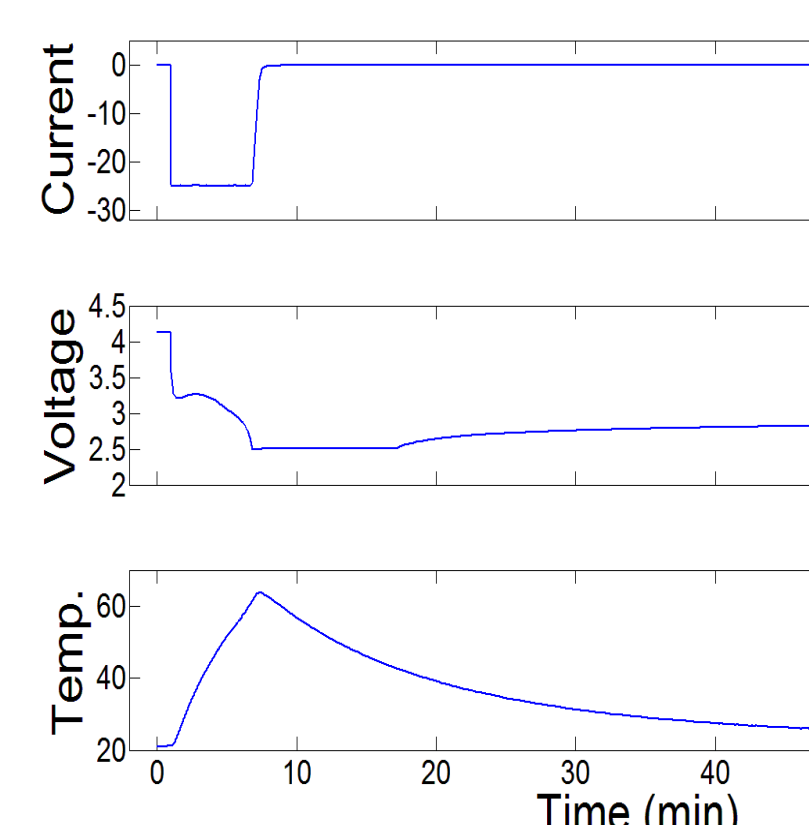
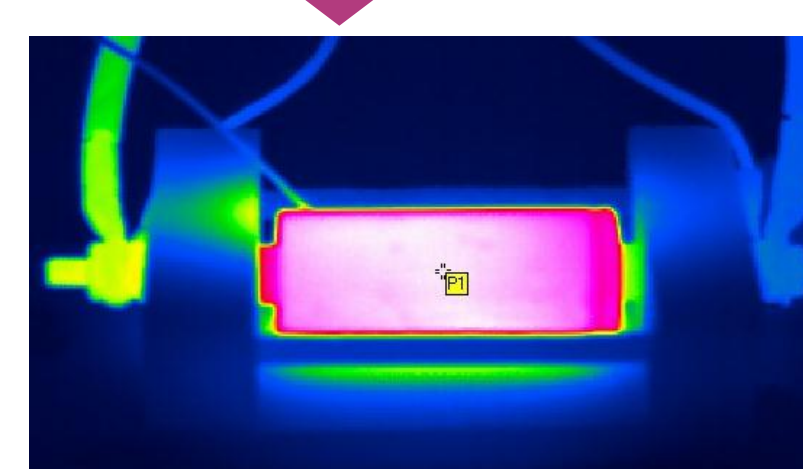
## Experiment

**General characterization**

cell type	weight [g]	volume [l]	VN [V]	C [Ah]	E [Wh]	spec C [Ah/kg]	C density [Ah/l]	spec E [Wh/kg]	E density [Wh/l]
A123 ANR26650	72,90	0,0345	3,3	2,45	7,75	33,5	70,9	106,2	224
Sony US26650VT	90,52	0,0345	3,7	2,67	9,57	29,5	77,4	105,7	277
Panasonic CGR-26650B	93,35	0,0345	3,6	3,36	11,9	36,0	97,4	127,8	346

**Battery cycling**

Operation characteristics of SONY US26650VT



Abuse experiments such as short circuit, nail penetration and overcharge will be conducted.